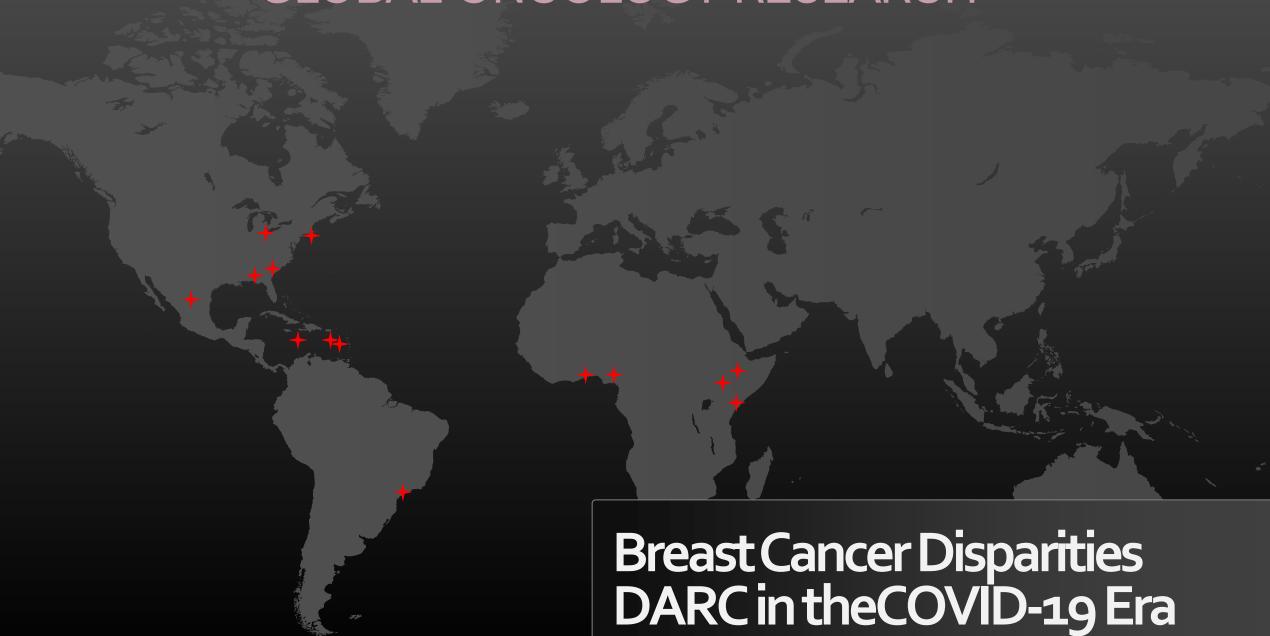
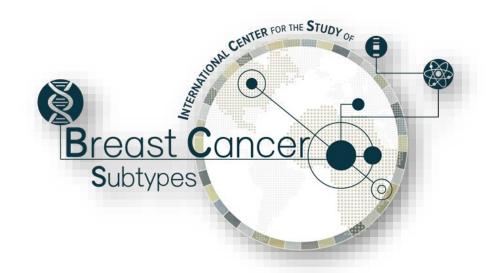


GLOBAL ONCOLOGY RESEARCH





I have no disclosures





Outline of "Stories"

Topics to Discuss

- Emergence of Disparities and Unmasking Tumor Biology
- Population Private Risk Alleles
 - Previous race-associated risk
 - BRCA1/2 in Africans
- Race vs Ancestry in Gene Network Changes
 - Cancer Pathways (altered)
 - Immune response (altered)
- Tumor Immunology Evolution at Work
 - DARC phenotype?
 - Novel tumor cell
- COVID-19 Disparities the DARC connection (proposal)
- Impact of COVID-19 in cancer patients (emerging... and variable)

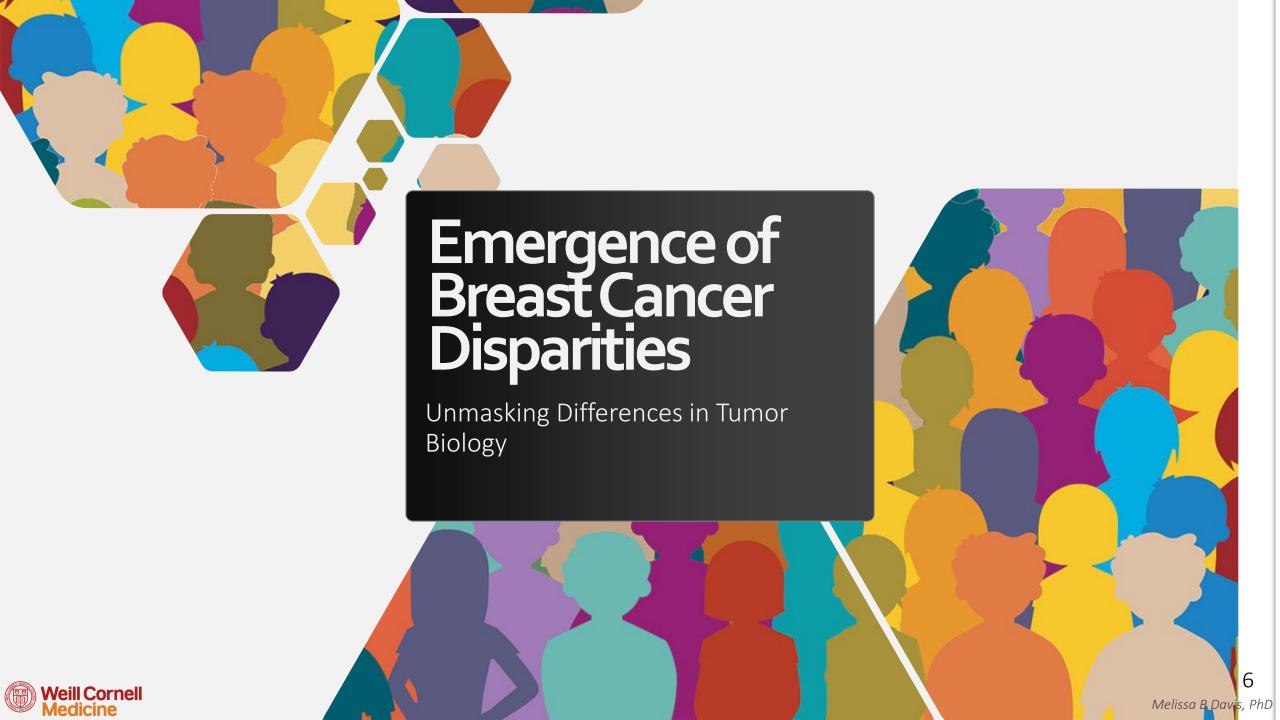




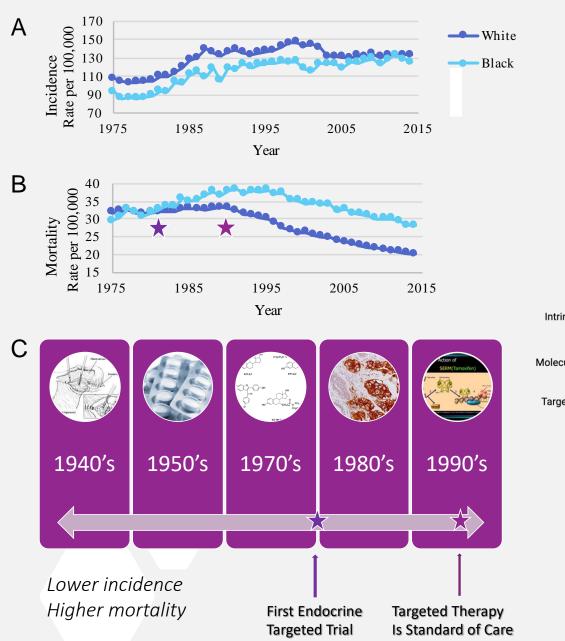
higher for African American women in the U.S. than Caucasian women.

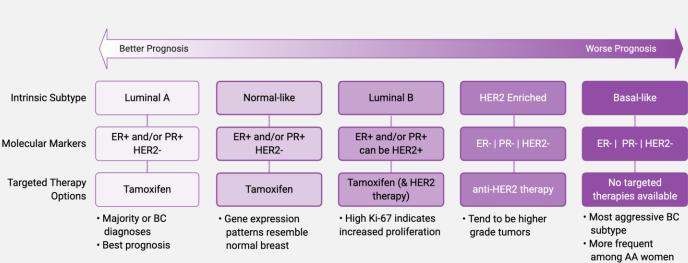
Background of Breast Cancer Disparities

Emergence of biased outcomes



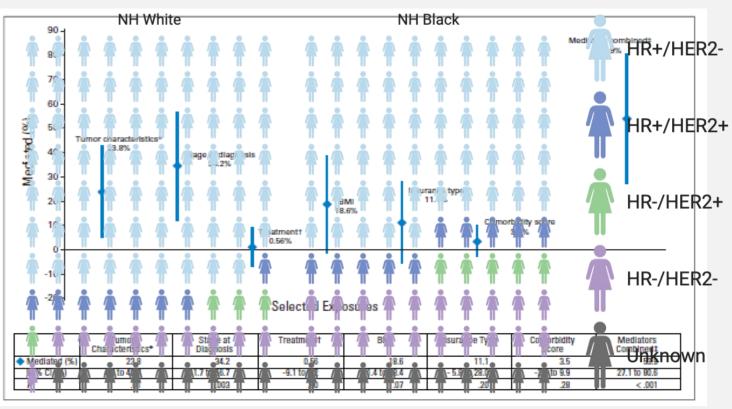
Emergence of Survival Disparity







Racial Differences in Survival – Contemporary Mediating Effectors



Warner ET, Tamimi RM, Hughes ME, et al. Racial and Ethnic Differences in Breast Cancer Survival: Mediating Effect of Tumor Characteristics and Sociodemographic and Treatment Factors. J Clin Oncol. 2015;33(20):2254–22 istribution Job BO Subtypes among non-Hispanic White and non-Hispanic Black women in the US, 2012-2016. Data from the SEER program and CDC National Program of Cancer Registries, statistics reported from DeSantis et al 2019. (Breast Cancer Statistics, 2019)



KNOW THE SCIENCE



HR = Hormone receptor

HR+ means tumor cells have receptors for the hormones estrogen or progesterone, which can promote the growth of HR+ tumors. Hormone therapies like tamoxifen can be used to treat HR+ tumors.

HER2 = Human epidermal growth factor receptor

HER2+ means tumor cells overexpress (make high levels of) a protein, called HER2/neu, which has been shown to be associated with certain aggressive types of breast cancer. Trastuzumab and some other therapies can target cells that overexpress HER2.

HR+/HER2-

····· aka "Luminal A"

73% of all breast cancer cases

- Best prognosis
- Most common subtype for every race, age, and poverty level



HR-/HER2-

..... aka "Triple Negative"

13% of all breast cancer cases

- Worst prognosis
- Non-Hispanic blacks have highest rate of this subtype at every age and poverty level



HR+/HER2+

..... aka "Luminal B"

10% of all breast cancer cases

• Little geographic variation by state



HR-/HER2+

..... aka "HER2-enriched"

5% of all breast cancer cases

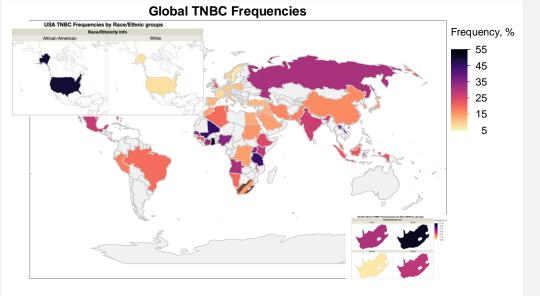
· Lowest rates for all races and ethnicities

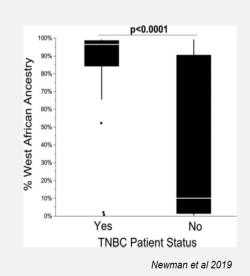


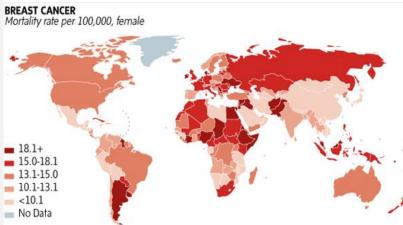


Global perspectives of Breast Cancer Incidence and Mortality



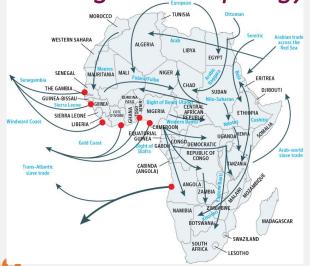


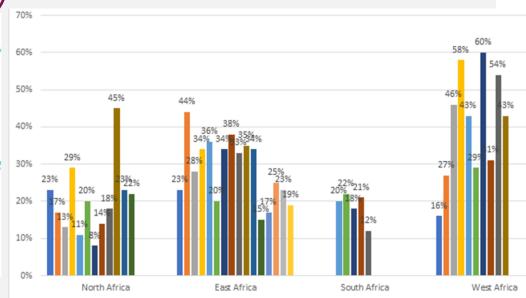




Africa has lowest incidence, but highest mortality burden *Adapted from WHO data*



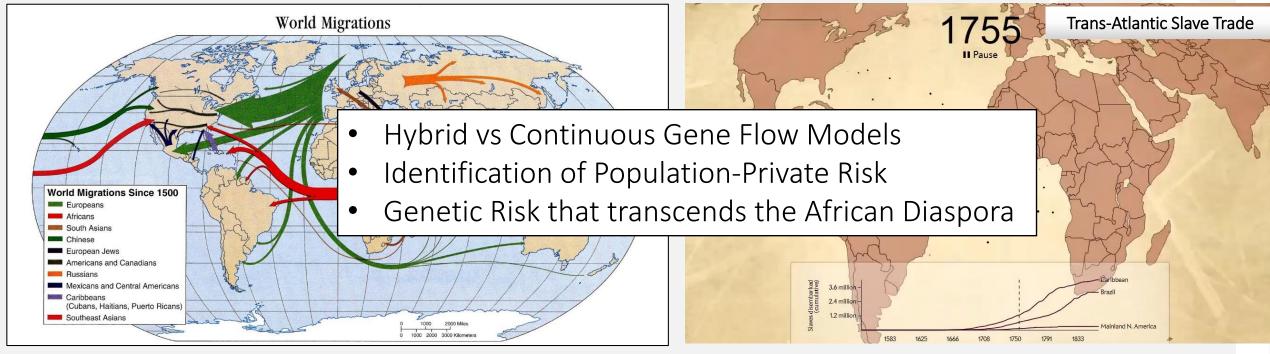


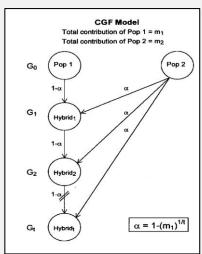






Global Human Populations and African Ancestry

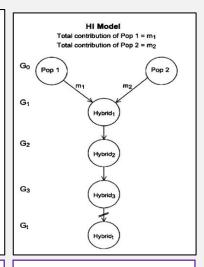




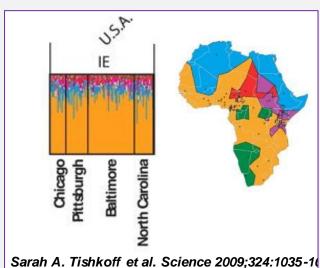
Influx of parental

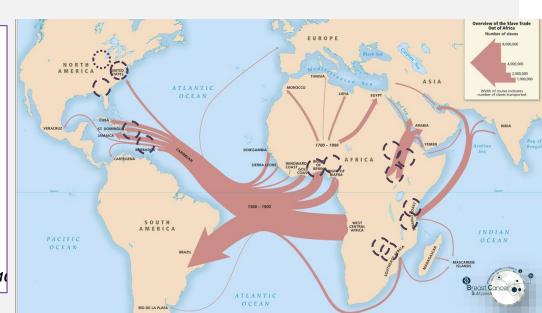
(African) alleles

Medicine

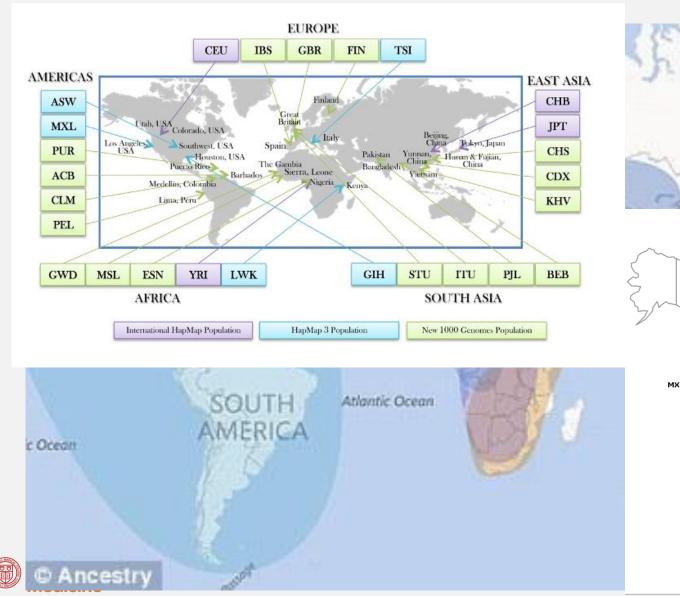


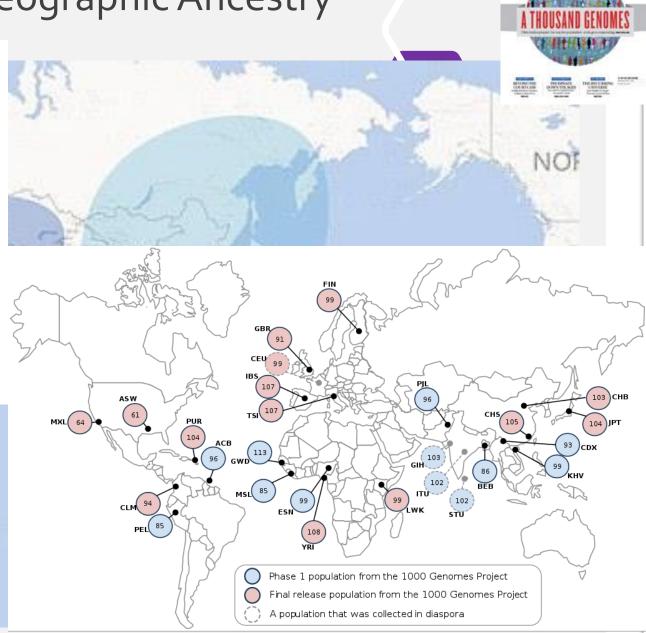
Recombination of the founder alleles + drift





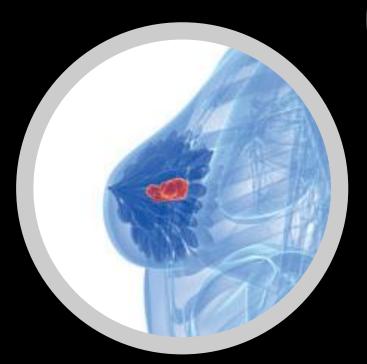
Ancestrally Informative Markers Geographic Ancestry

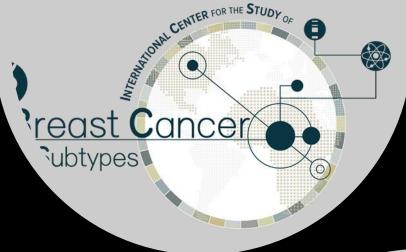




nature



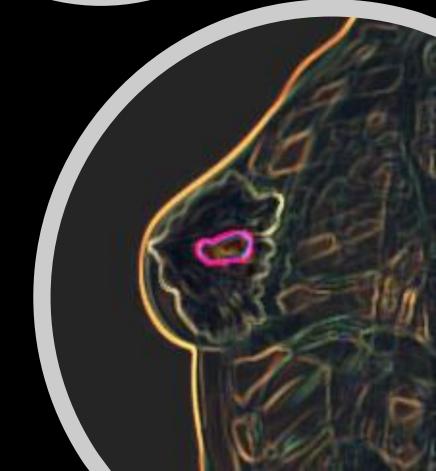




Differences in TNBC Tumor Biology

Emerging Evidence





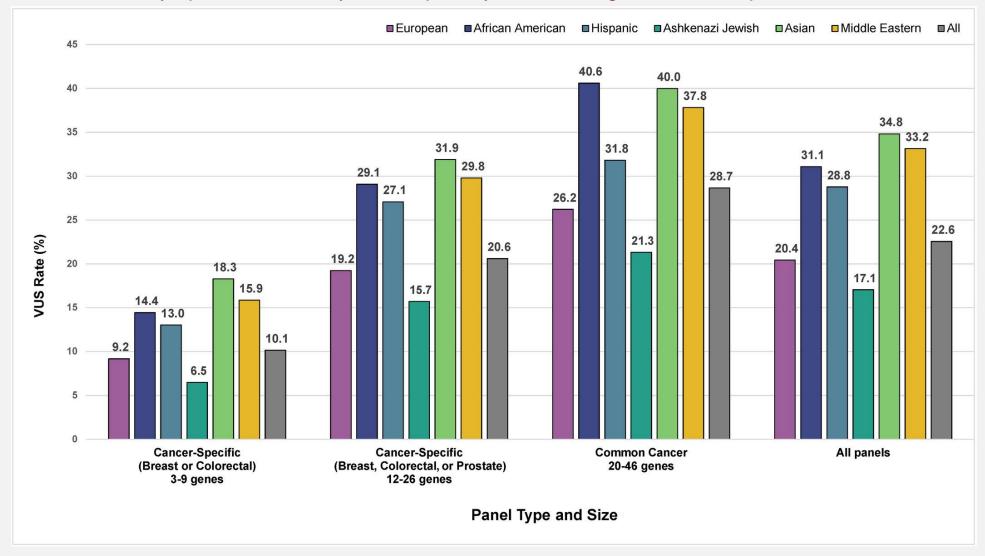


Population-private Risk Alleles

Power of Inclusion

Melissa B Davis, PhD 17

Ancestry-specific hereditary cancer panel yields: Moving toward more personalized risk



Jrnl of Gene Coun, Volume: 29, Issue: 4, Pages: 598-606, First published: 30 March 2020, DOI: (10.1002/jgc4.1257)



ASA PAPER

Hereditary Susceptibility for Triple Negative Breast Cancer Associated With Western Sub-Saharan African Ancestry

Results From an International Surgical Breast Cancer Collaborative

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Dial Vittles DhD | and Maliana D Davis DhD*

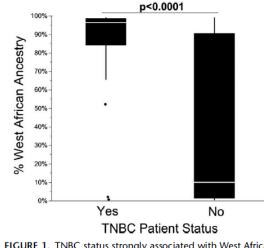


FIGURE 1. TNBC status strongly associated with West African ancestry. TNBC patient status compared to percent West African ancestry (n = 82; TNBC yes = 44, TNBC no = 38; Student t test, P < 0.0001, CI [-0.6078, -0.3066]).

Models With Covariates*

	Minor Allele	Sample Size	By SNV Genotypes Weighted (Dominant Model)				By SNV Alleles Weighted (Dosage model)					
SNV ID			OR	SE	Lower 95% CI	Upper 95% CI	P Value	OR2	SE2	Lower 95% CI2	Upper 95% CI2	P Value
rs2814778	С	297	0.283	0.896	0.049	1.637	0.0473	0.782	0.346	0.397	1.54	0.000345
rs13000023	A	50	0.077	1.187	0.008	0.790	0.0308	0.258	0.691	0.067	0.9998	0.04996
rs2363956	G	43	0.106	1.271	0.009	1.278	0.0773	0.132	0.955	0.020	0.855	0.03369
rs2981578	C	50	0.778	1.215	0.072	8.415	0.8362	0.778	1.215	0.072	8.415	0.8362
rs2981579	G	50	1.121	1.254	0.096	13.090	0.9271	0.882	0.727	0.212	3.669	0.8631
rs3112572	A	44	0.746	0.895	0.129	4.313	0.7438	0.941	0.764	0.210	4.205	0.936
rs3745185	A	45	0.813	1.014	0.111	5.932	0.8378	0.813	1.014	0.111	5.932	0.8378
rs4245739	C	49	0.718	0.869	0.131	3.944	0.7032	0.469	0.632	0.136	1.620	0.2315
rs4849887	T	48	6.725	1.207	0.631	71.680	0.1145	3.057	0.862	0.564	16.560	0.1949
rs609275	C	50	0.562	0.931	0.091	3.483	0.5357	0.562	0.931	0.0907	3.483	0.5357

^{*}rs2814778 was adjusted for race and age while all other SNVs were adjusted for race.

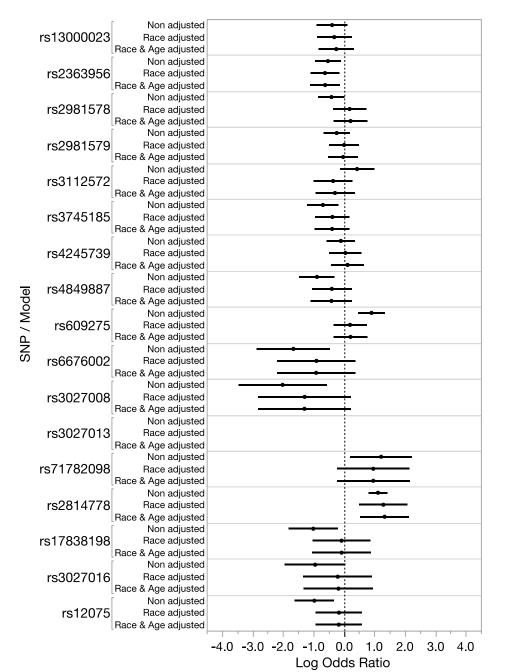


CI indicates confidence interval; OR, odds ratio; SE, standard error; SNV, single-nucleotide variant.

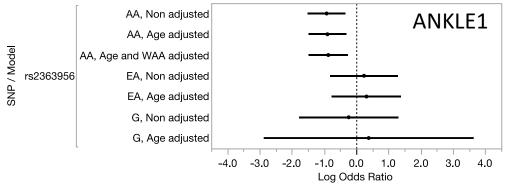
Dominant Model: evaluates association between homozygous minor allele status and breast cancer risk.

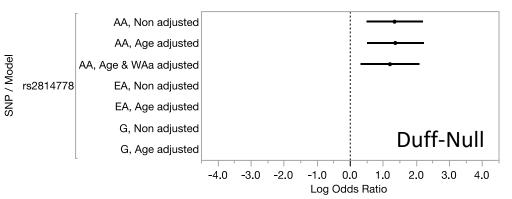
Dosage Model: evaluates association between presence of the minor allele (whether homozygous or heterozygous) and breast cancer risk.

TNBC case-series risk analysis of SNVs previously identified as BC risk alleles

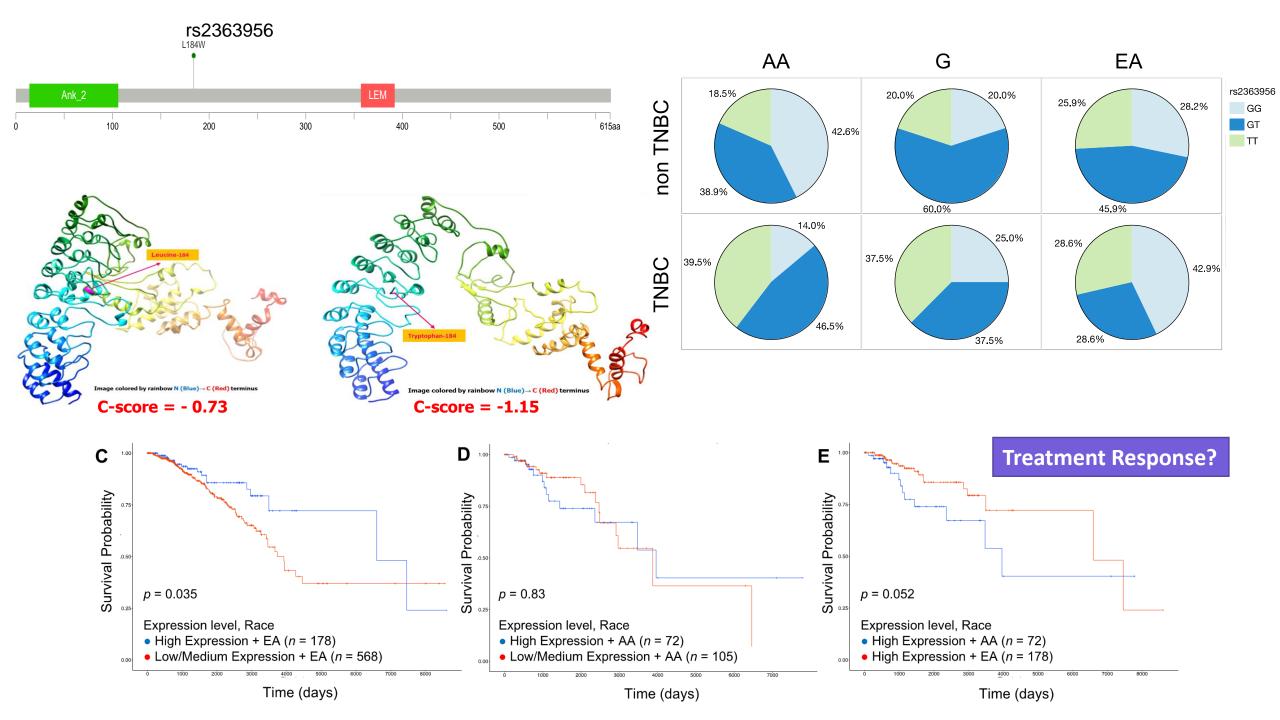


Variant	Region	Associated/Neighboring Gene(s)
rs2814778	1q23	DARC/ACKR1
rs13000023	2q35	TNP1, DIRC3
rs2363956	19p13.11	ANKLE1
rs2981578	10q26	FGFR2
rs2981579	10q26.13	FGFR2
rs3112572	16q12	LOC643714
rs3745185	19p13	BABAM1
rs4245739	1q32.1	MDM4
rs4849887	2q14.2	LOC84931, GLI2
rs609275	1 Îq13	MYEOV,CCND1



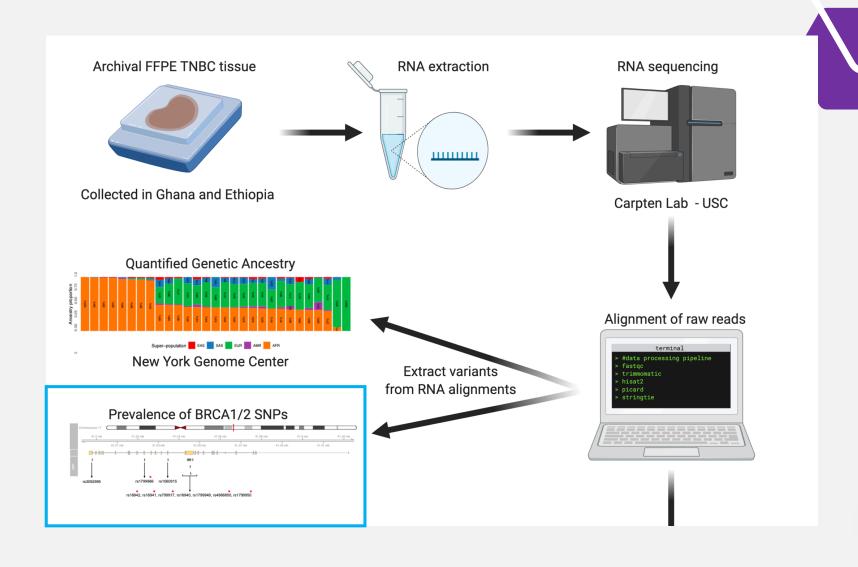








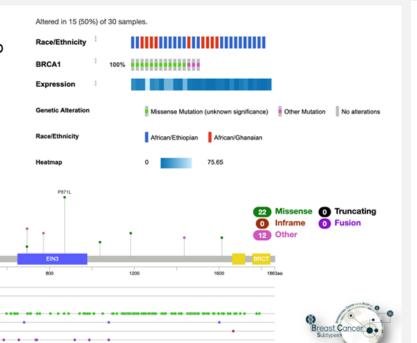
Premise: Taking SNP calls used for ancestry quantification, can we detect BRCA1/2 [somatic] mutations in our African and domestic RNAseq TNBC cohorts?

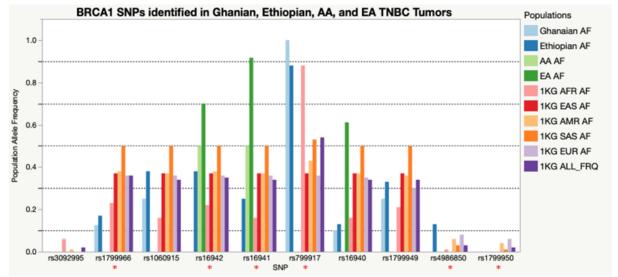


Breast Cancel



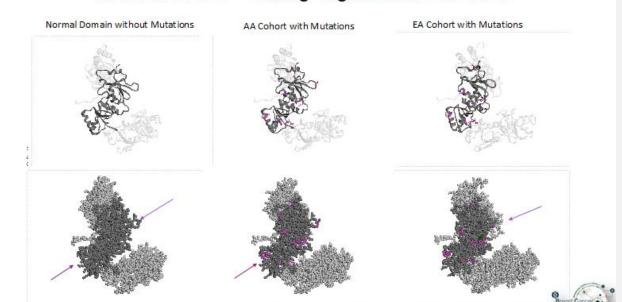
BRCA1 SNPs lollipop plot, domain structure, and heatmap for Ethiopian and Ghanaian cohorts





BRCA1 SNPs in Ghanaian, Ethiopian, AA, and EA TNBC tumors compared against 1000 Genome super group populations, ordered by CHR position

BRCA1 SNPs - Giving 'significance' to VUS



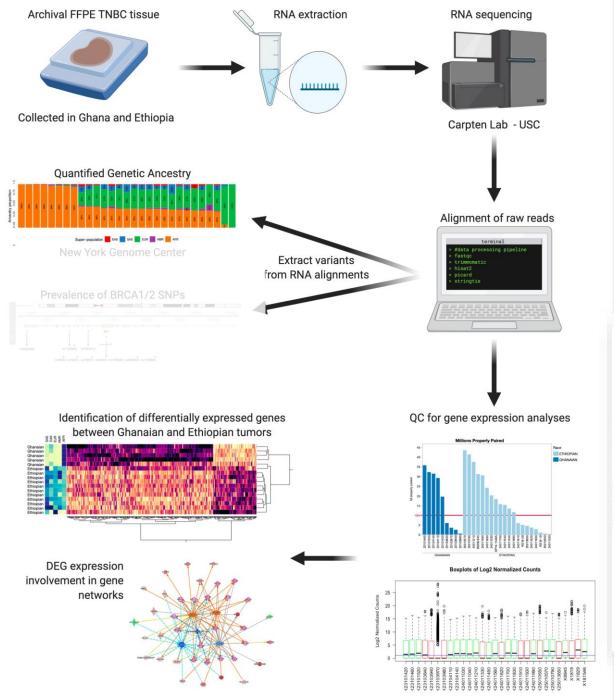


Cancer Hotspots
 OncoKB
 PTM Sites €

BRCA2 SNPs identified in Ghanaian and Ethiopian TNBC tumors (from RNAseq)

SNP rsID	Variant Type	AA Change	SIFT / Polyphen	Clinical Significance	Ghana	aian AF	Ethiop	oian AF	1KG AFR AF	1KG Global AF
rs1799943	5 Prime UTR			Benign	0.00	n = 1	-	-	0.23	0.21
rs766173	Missense	N289H	Deleterious / Possibly Damaging	Likely benign / Conflicting Interpretations of Pathogenicity	-	-	0.00	n = 3	0.07	0.07
rs144848	Missense	N372H	Tolerated / Possibly Damaging	Benign / Uncertain significance	0.13	n = 4	0.30	n = 7	0.31	0.25
rs1801439	Synonymous	S455S		Benign	0.00	n = 2	0.00	n = 4	0.07	0.07
rs1801499	Synonymous	H743H		Benign	0.00	n = 5	0.00	<i>n</i> = 9	0.07	0.07
rs1799944	Missense	N991D	Tolerated / Benign	Benign	0.00	n = 1	0.00	n = 2	0.07	0.08
rs1801406	Synonymous	K1132K			0.00	n = 3	0.10	n = 14	0.27	0.27
rs543304	Synonymous	V1269V		Likely benign	0.00	n = 2	0.00	n = 4	0.17	0.17
rs206075	Synonymous	L4563L		Likely benign	1.00	n = 5	1.00	n = 3	1.00	0.97
rs1799955	Synonymous	S2414S		Likely benign	0.00	n = 1	0.40	n = 4	0.22	0.23
rs169547	Missense	V2466A	Tolerated / Benign	Uncertain significance	-	-	1.00	n = 1	1.00	0.98
rs9590940	Synonymous	V2820V		Likely benign	0.00	n = 5	0.00	n = 5	0.00	0.01
rs1801426	Downstream / Missense	13412V	Tolerated / Benign	Uncertain significance	0.00	n = 1	0.30	n = 2	0.02	0.04





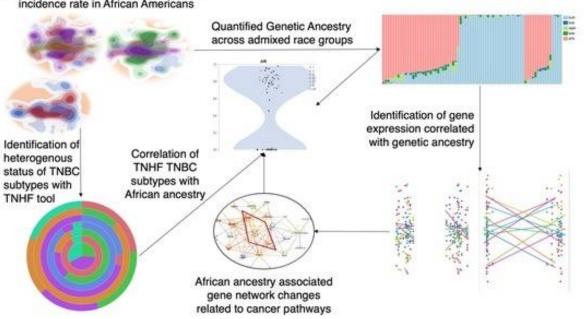


Identification of Distinct Heterogenic Subtypes and Molecular Signatures Associated with African Ancestry in Triple Negative Breast Cancer Using Quantified Genetic Ancestry Models in Admixed Race Populations

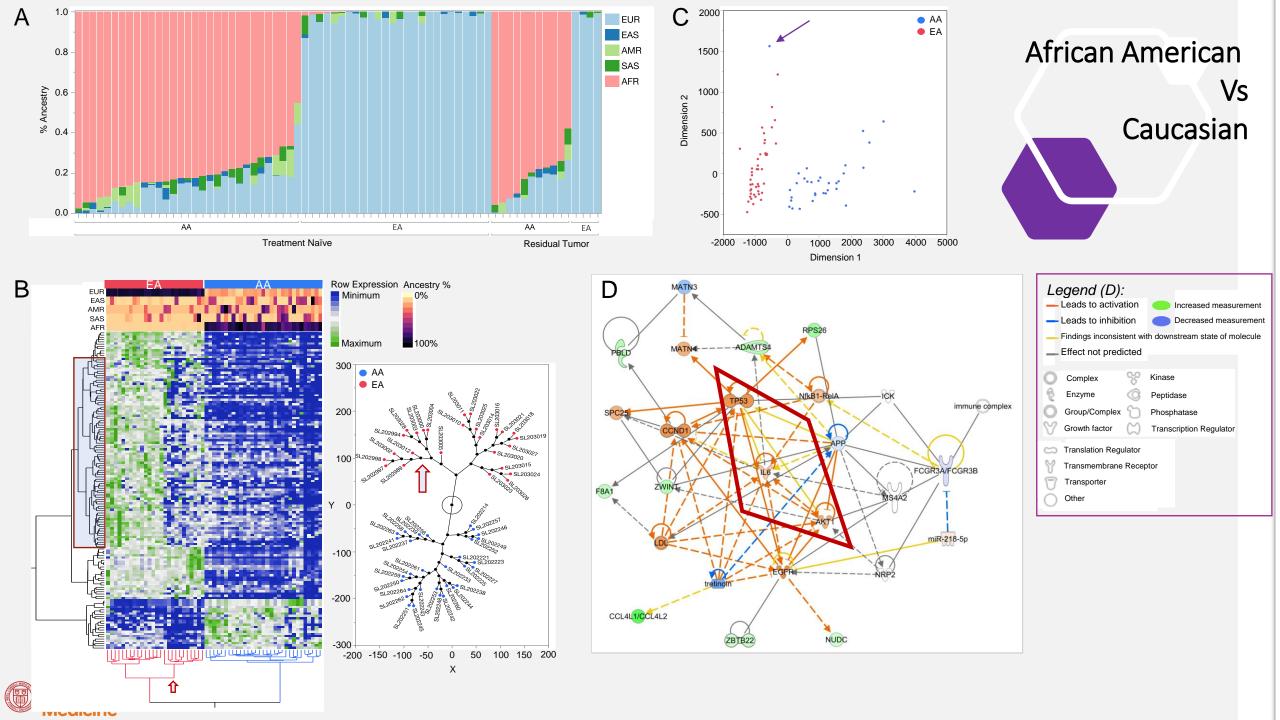
by Melissa Davis 1.† 🗵 5, Rachel Martini 1.† 🗵 1. Lisa Newman 1 🗷 1. Olivier Elemento 2.3.4 🗵 1. Jason White 5 🗷 1. Akanksha Verma 6 🖾 1. Indrani Datta 7 🖾 1. Indrani Datta 7 🖂 1. Indrani Datta 9 1. Indrani

- Department of Surgery, Weill Cornell Medicine, New York, NY 10065, USA
- Department of Physiology and Biophysics, Weill Cornell Medicine, New York, NY 10065, USA
- ³ Caryl and Israel Englander Institute for Precision Medicine, Weill Cornell Medicine, New York, NY 10065, USA
- Institute for Computational Biomedicine, Weill Cornell Medicine, New York, NY 10065, USA
- Department of Biology and Center for Cancer Research, Tuskegee University, Tuskegee, AL 36088, USA
- Department of Computational Biology, Weill Cornell Medicine, New York, NY 10065, USA
- Department of Public Health Sciences, Henry Ford Health System, Detroit, MI 48202, USA

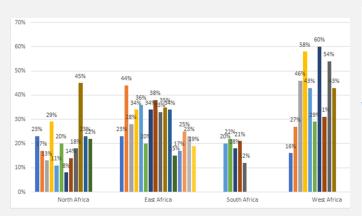
Heterogeneous TNBC tumors with higher incidence rate in African Americans

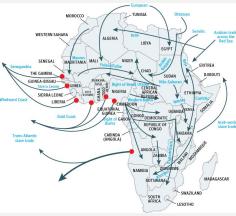






Pan-African Comparisons



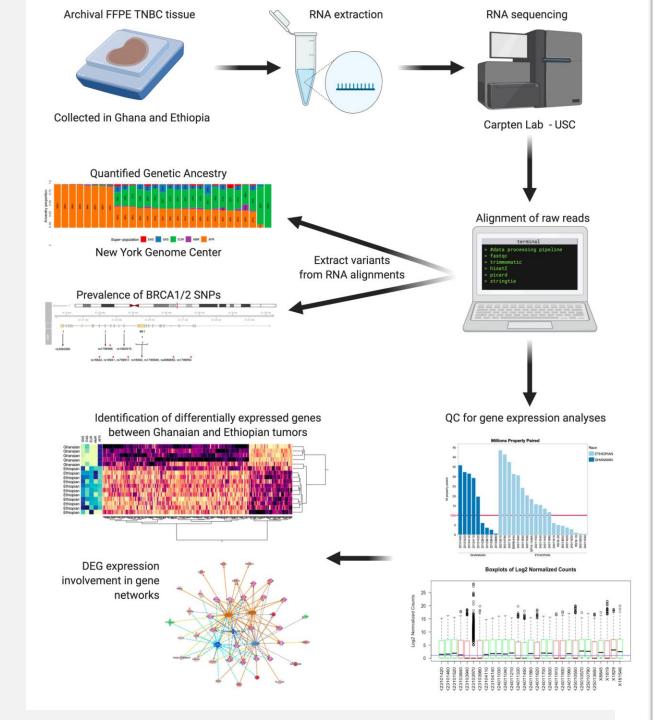


RNA sequencing of TNBC tumors from Ghanaian and Ethiopian women

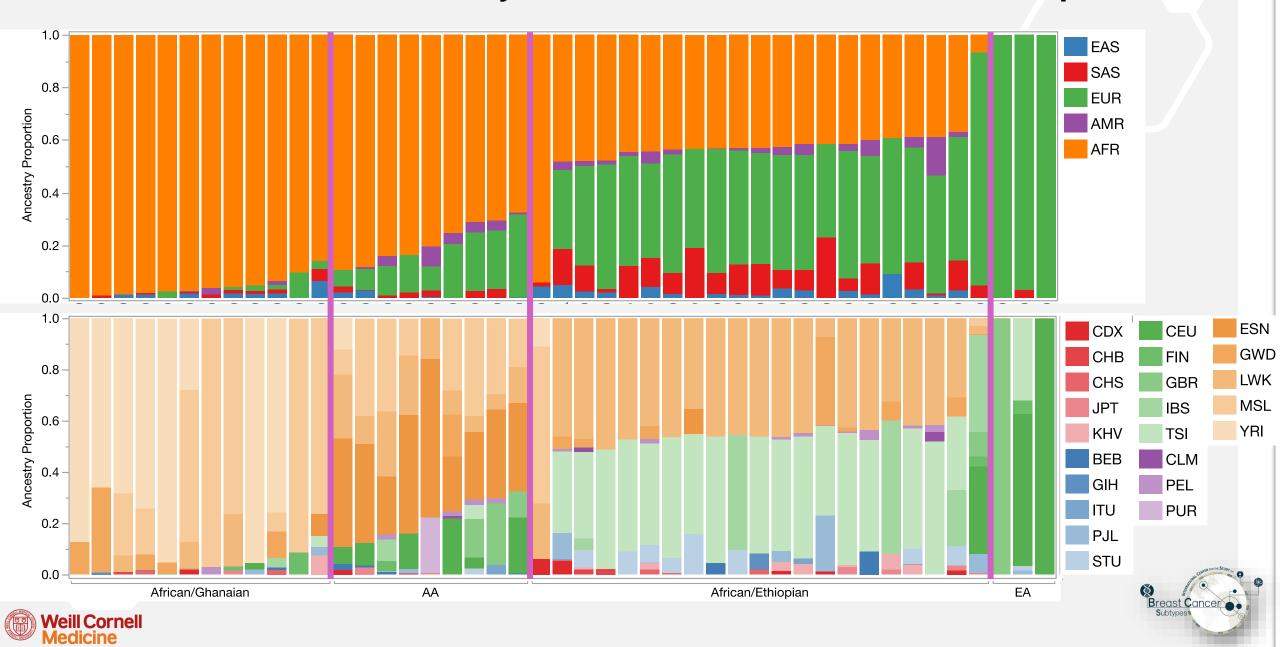
- 1. Ancestry Gene Network Function
- 2. Novel mutations in Cancer Genes



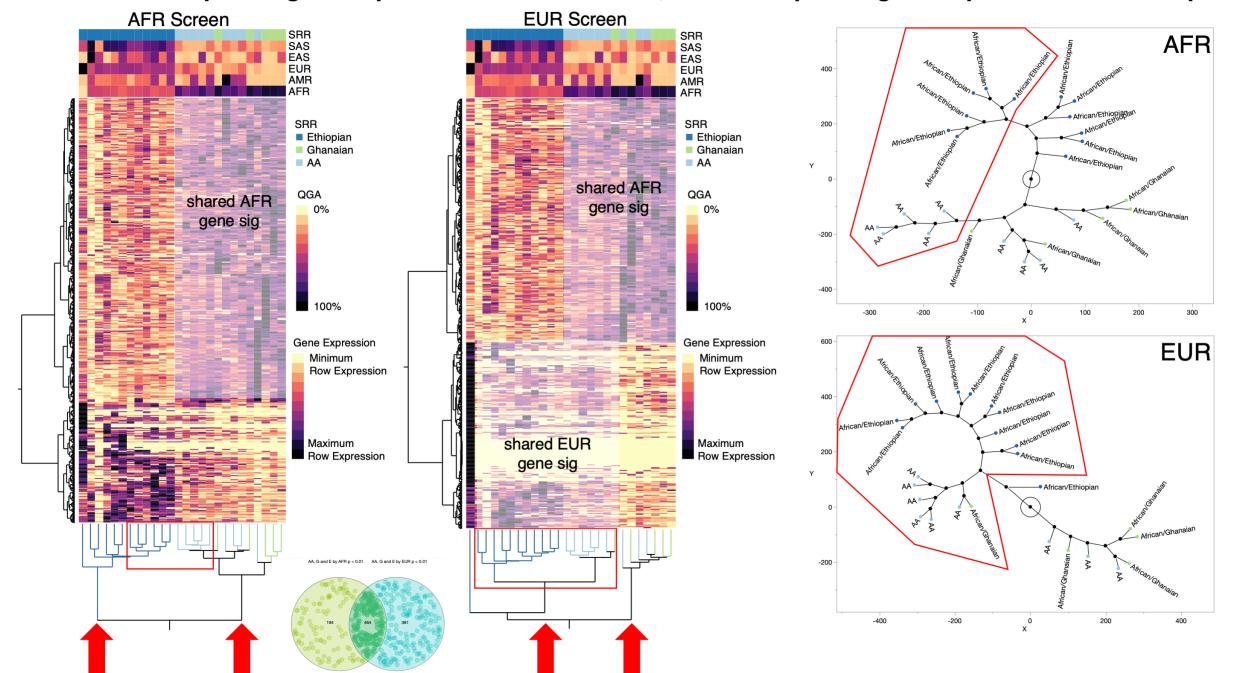




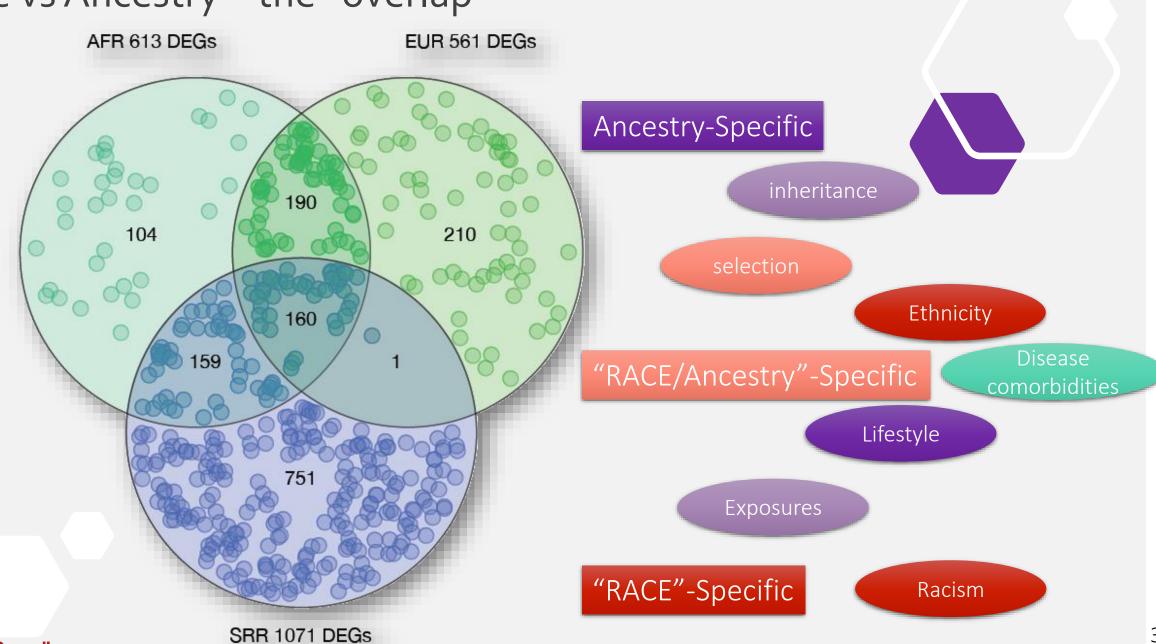
Quantified Genetic Ancestry Estimations in our ICSBCS RNAseq Cohort



AAs share AFR-specific gene expression with Ghanaians, and EUR-specific gene expression with Ethiopians

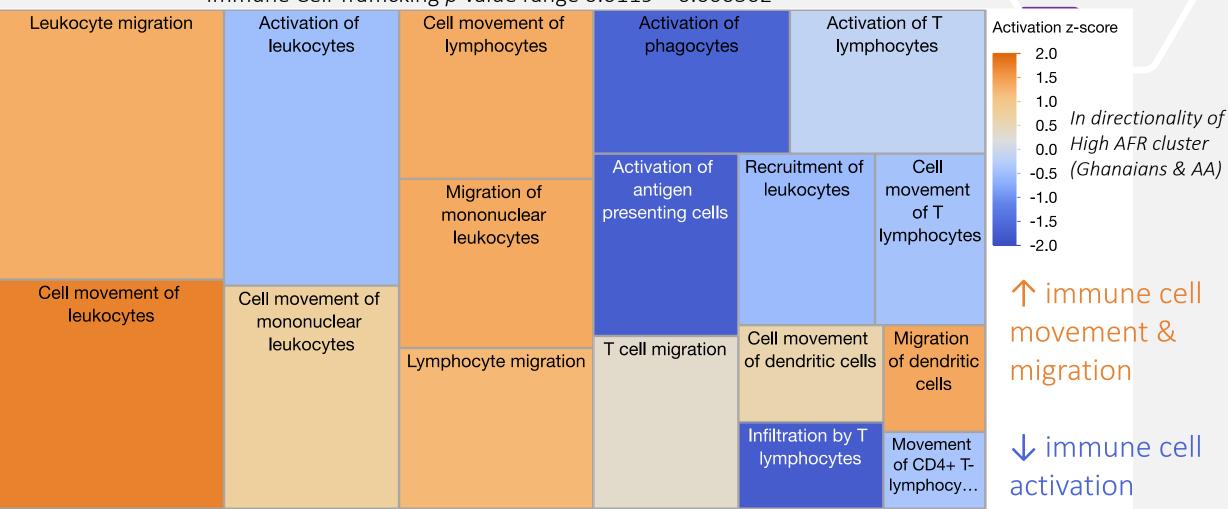


Race vs Ancestry – the "overlap"



Increase in immune cell movement and migration, but not activation, associated with High AFR cluster genes

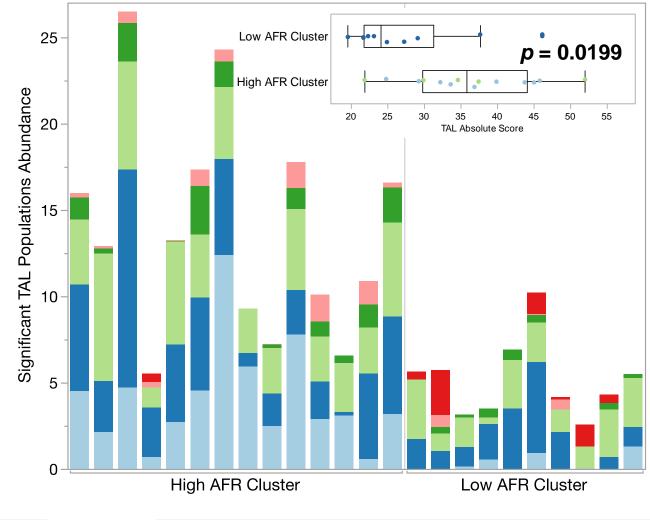
Immune Cell Trafficking *p* value range 0.0119 – 0.000502



Diseases or Functions Annotation



Abundance of tumor-associated leukocytes (TAL) significantly higher among High AFR cluster



B cells naive
T cells CD8
T cells follicular helper
NK cells activated
Monocytes
Mast cells activated

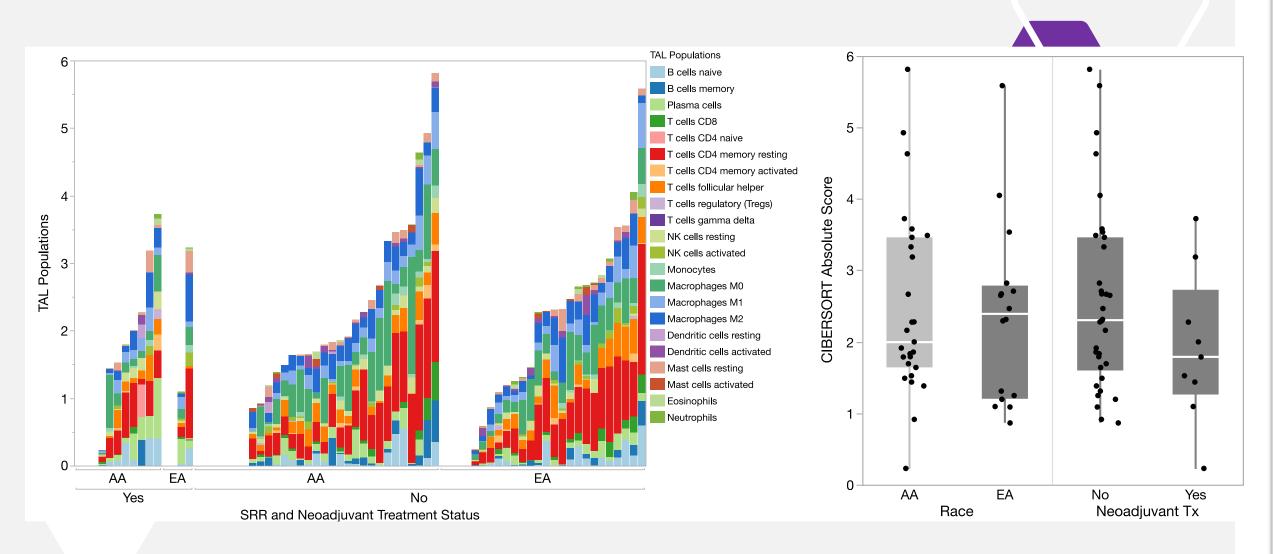


	TAL population	High AFR cluster mean	Low AFR cluster mean	Student's t- test <i>p</i> value	<i>ANOVA</i> by SRR <i>p</i> value
	B cells naive	4.12	0.29	0.0004	0.0016
	T cells CD8	4.16	1.87	0.0268	0.0248
	T cells follicular helper	3.98	1.98	0.0019	0.0037
	NK cells activated	1.01	0.27	0.0103	0.0455
	Monocytes	0.55	0.14	0.0291	0.0403
	Mast cells activated	0.02	0.61	0.0613	0.0665

Deconvolution using CIBERSORTx



AA vs EA TNBC Immune Profiles

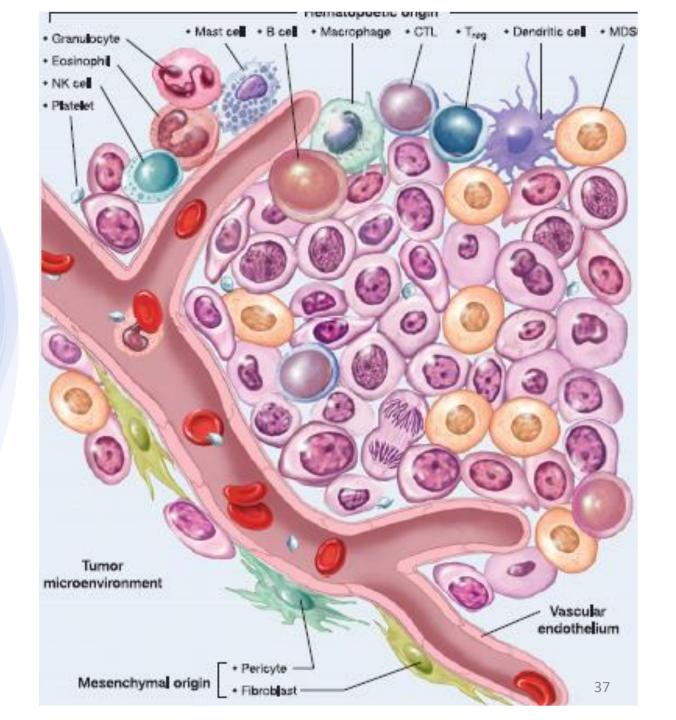


Davis, M.; Martini, R.; Newman, L.; Elemento, O.; White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, U.; Yates, C. White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, U.; Yates, C. White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, U.; Yates, C. White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, U.; Yates, C. White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, U.; Yates, C. White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, U.; Yates, C. White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, U.; Yates, C. White, J.; Verma, A.; Datta, I.; Adrianto, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, I.; Chen, Y.; Gardner, K.; Kim, H.-G.; Colomb, W.D.; Eltoum, I.-E.; Frost, A.R.; Grizzle, W.E.; Sboner, A.; Manne, I.; Chen, Y.; Gardner, M.; Chen, Yates, A.; Chen, Yates, A.;

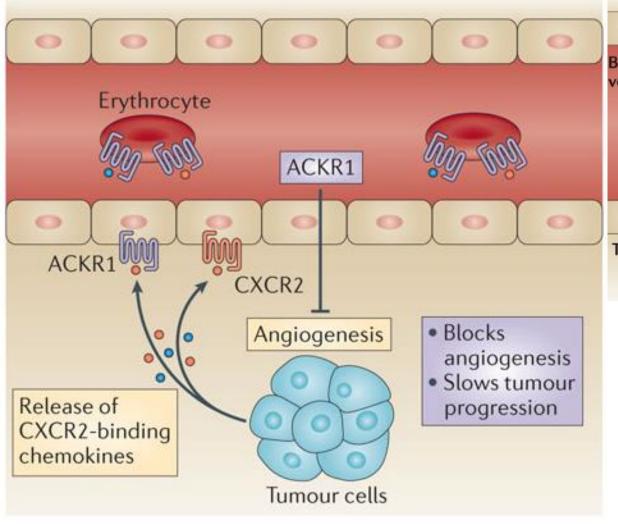


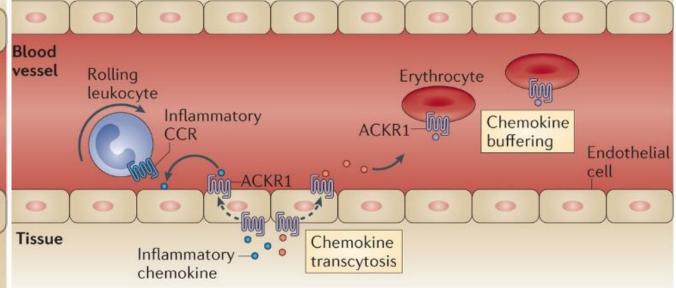
DARC Immune Tumor Phenotype

Unique immune response – ancestry specific

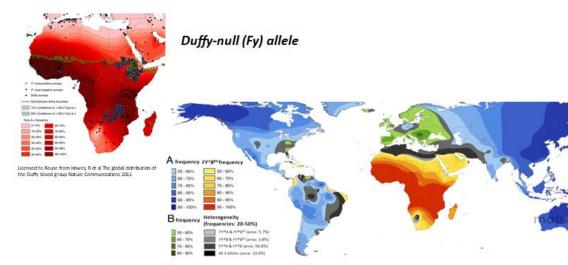


DARC regulates dynamic inflammatory chemokine levels via vascular endothelial cells and concentrations in plasma



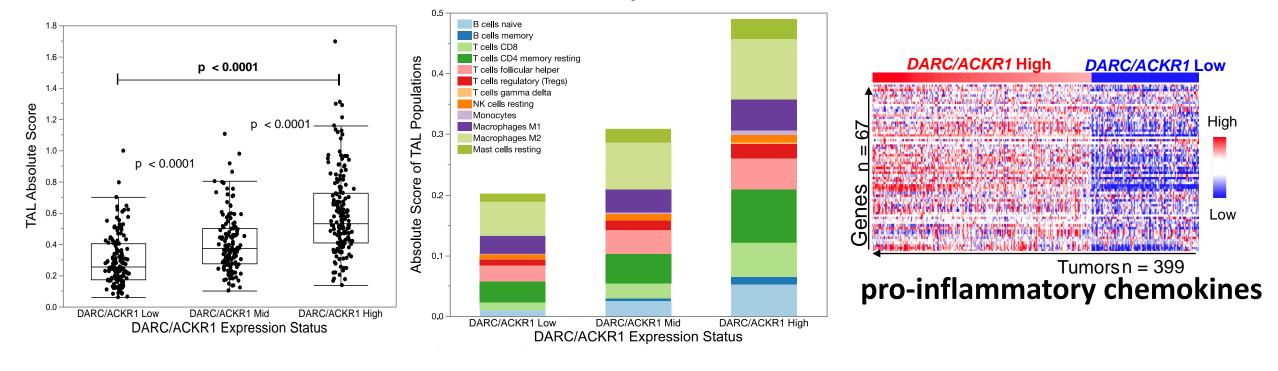


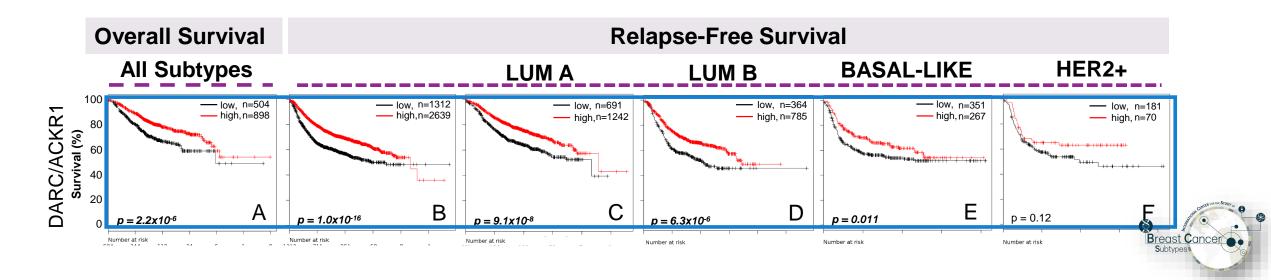
Genetic imprint of the West African Diaspora Impact on TNBC Risk?



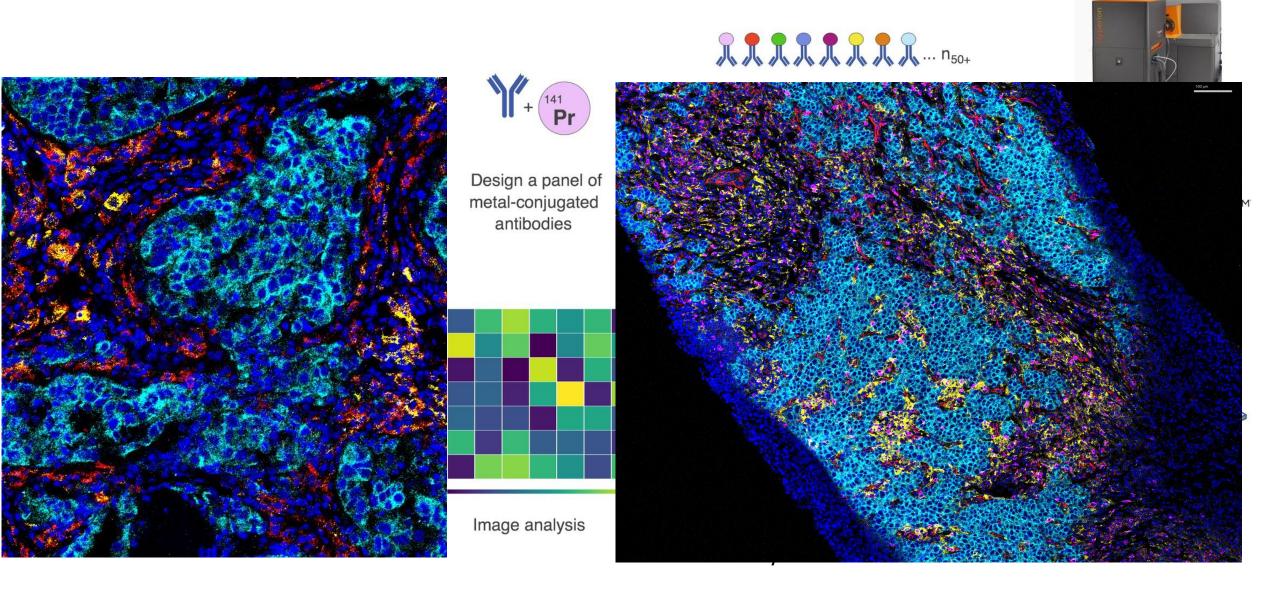
Nature Reviews | Immunology

DARC-associated tumor leukocyte associations in TCGA cohort





Imaging mass cytometry to define spatial distinctions Characterizing the DARC immuno-tumor-type





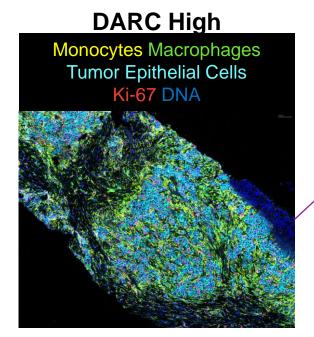
DARC high by IHC (n=2) 2x ROI from case = 4 ROI total DARC low by IHC (n=2) 2x ROI from case = 4 ROI total

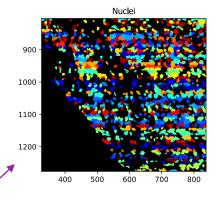
Hyperion™ IMC staining of ~30 markers

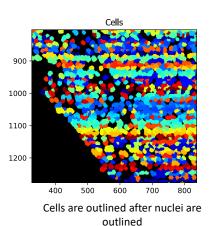
Cell segmentation
histoCAT™ single-cell protein analysis (meta - tSNE & phenograph clustering)

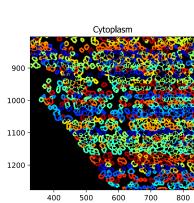
DARC Low

Tumor-associated Macrophages
Tumor Epithelial Cells
Monocytes DNA



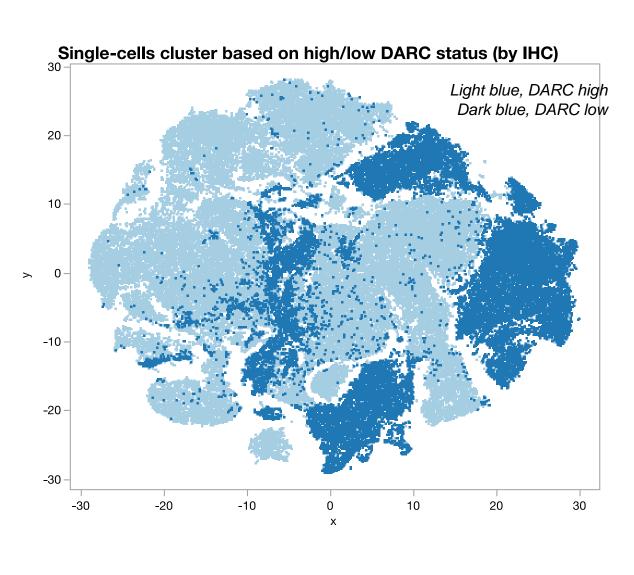


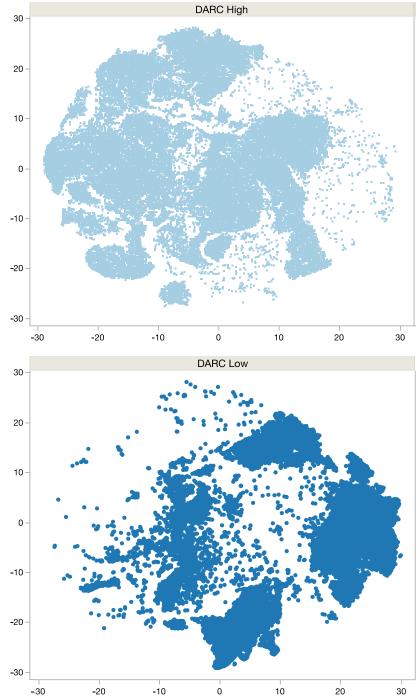


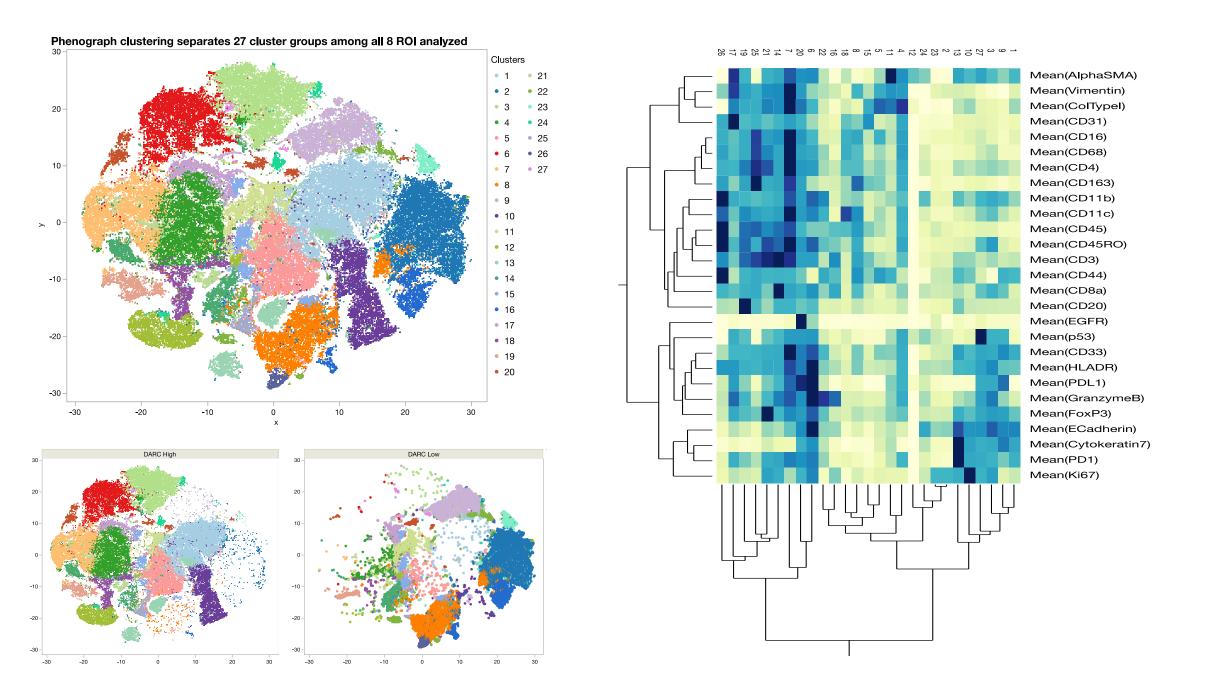


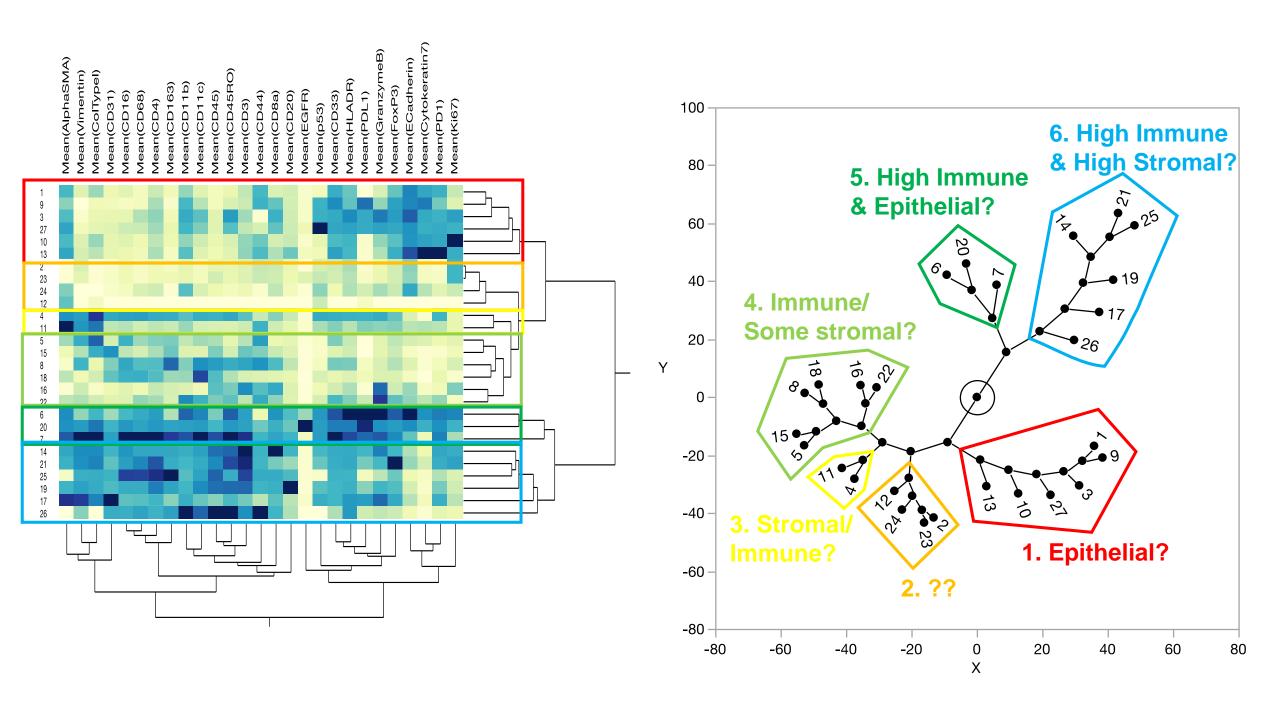
Cytoplasmic area is identified as the space b/t the nuclei and the cellular outline

DARC-specific clusters of tumor-associated cells



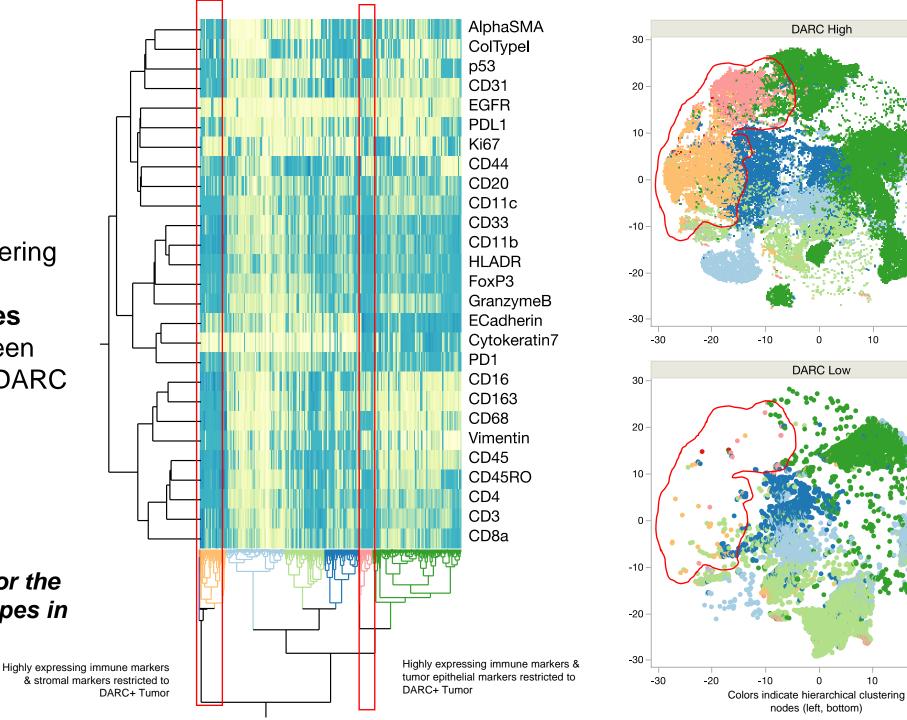






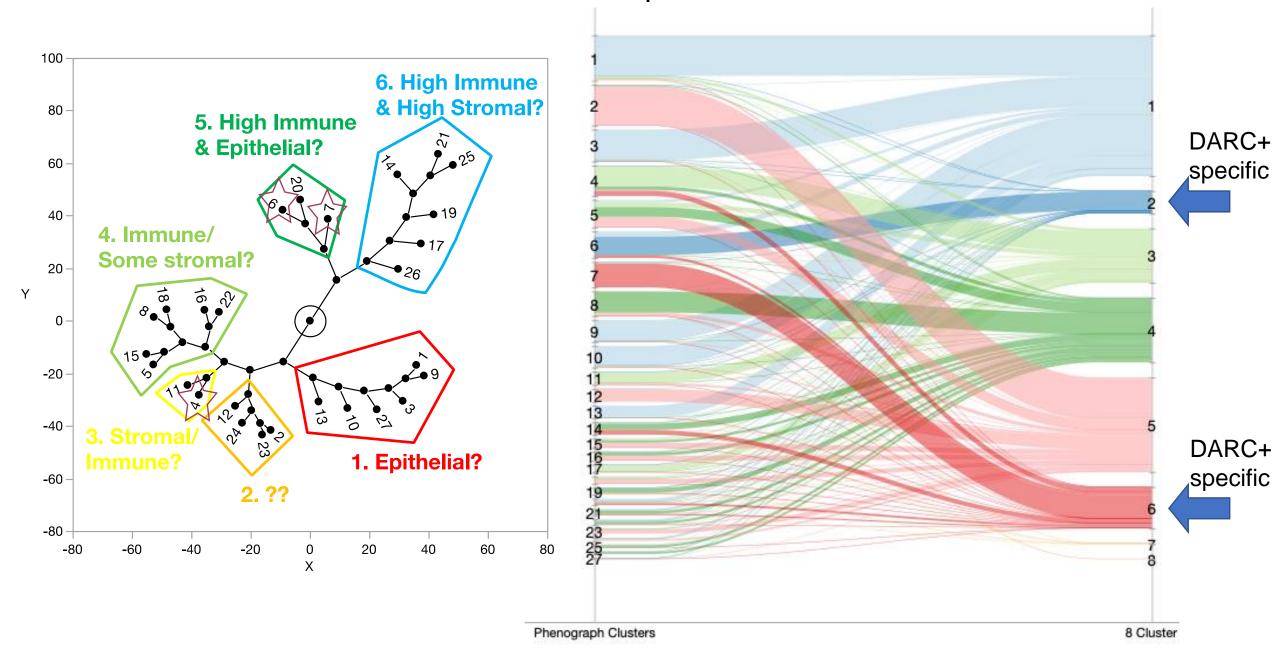
- Hierarchical clustering functionally descriptive nodes
- distinctions between DARC High and DARC low tumors
- 7 distinct nodes

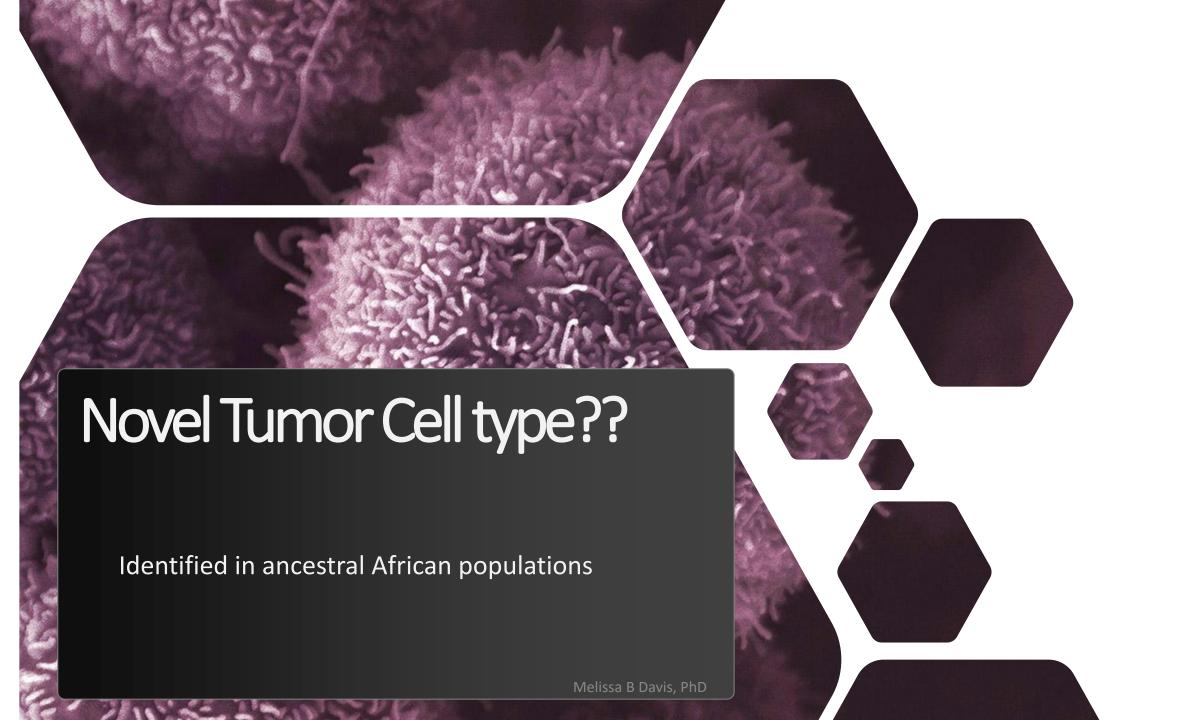
Next steps... Look for the 'new DARC phenotypes in tissues

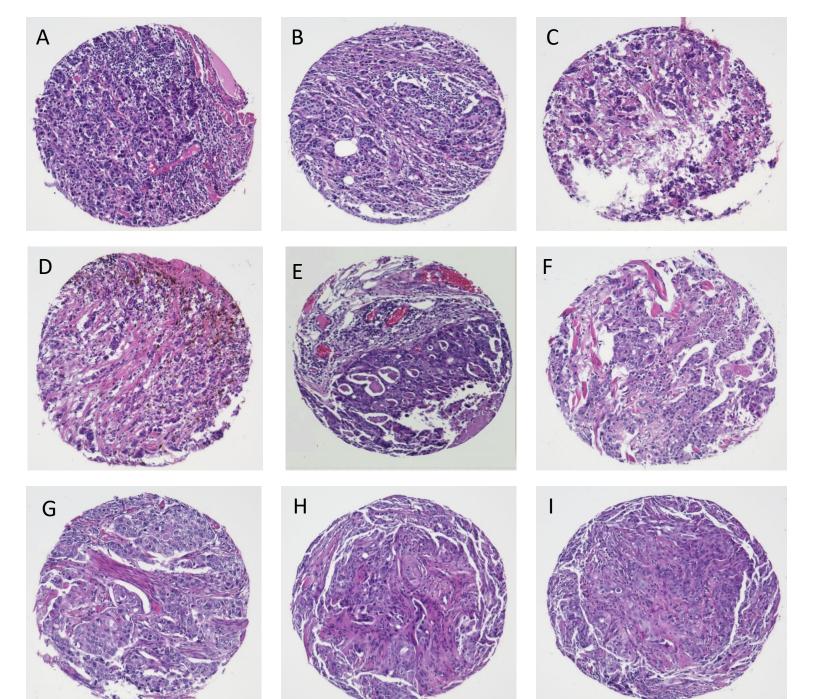


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Tumor-associated cells that are DARC+ specific?

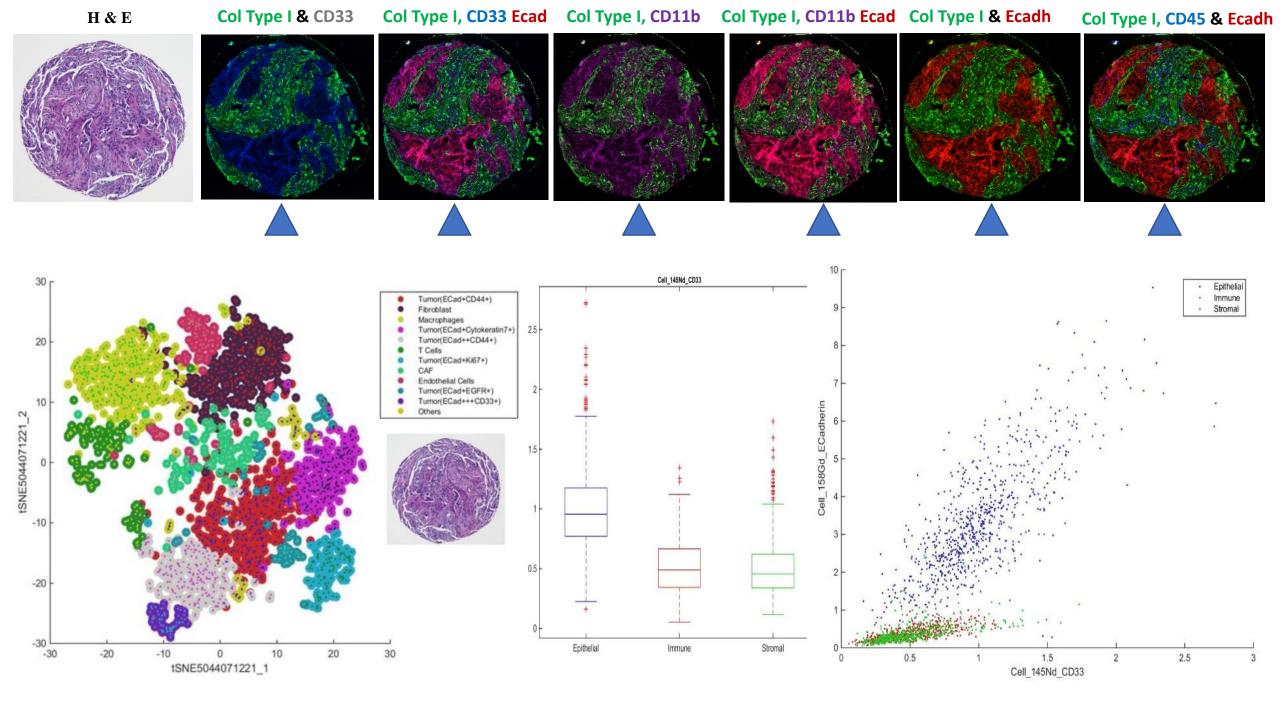






Ethiopian TMA

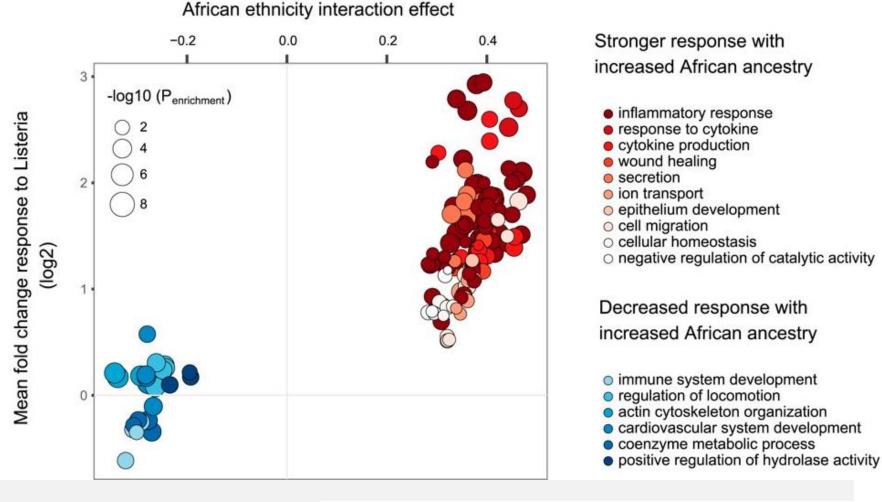
- Combination of Tumor Subtypes
- Investigation of tumor heterogeneity
- Distinctions associated with ancestry



Rationale: Distinct Immune Response – Evolution Influences

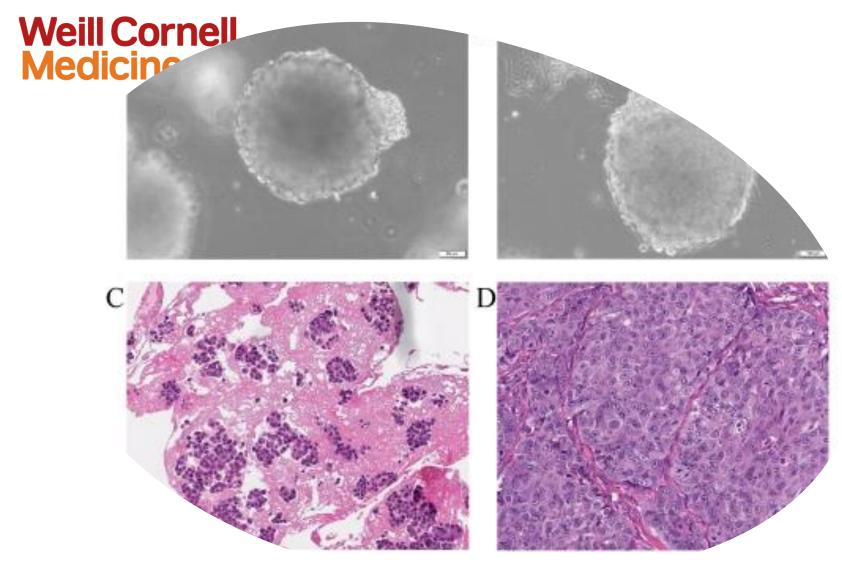
Human evolution genetics

- African Ancestry is associated with a stronger immune response
- Specific responses include wound healing, cytokine production and inflammatory response



Modified From: Nedelec, Y et. al, Genetic Ancestry and Natural Selection Drive Population Differences in Immune Responses to Pathogens. Cell, 2016. 167(3): p. 657-669 e21.





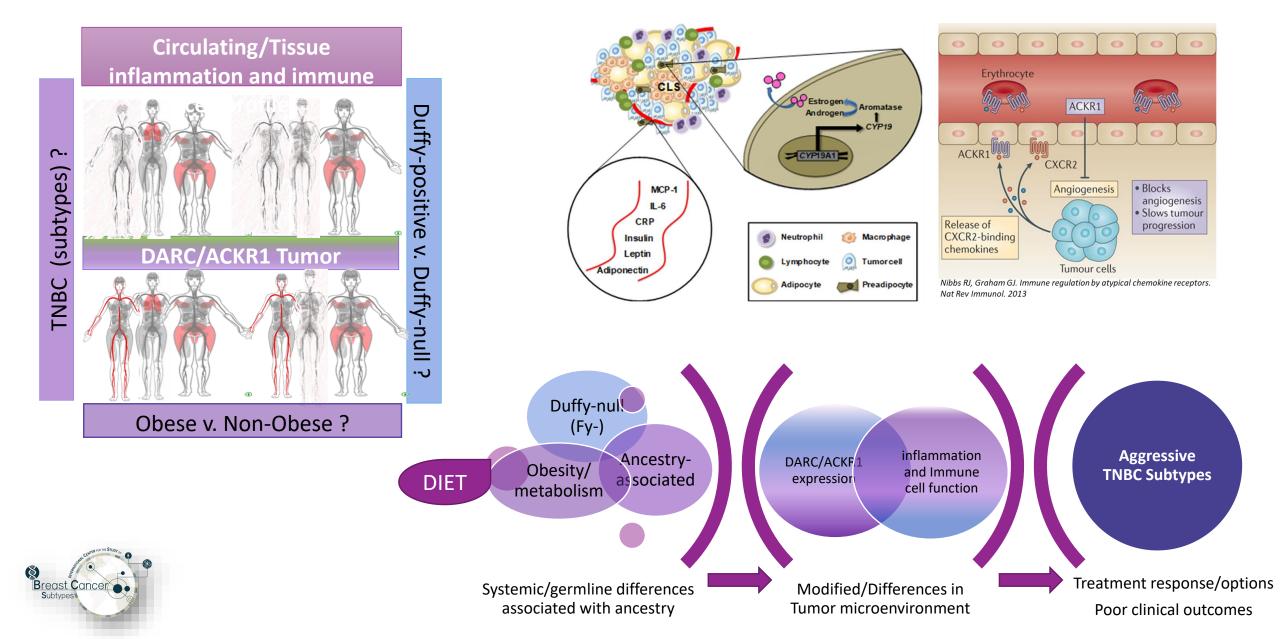
Future Directions Breast Cancer Disparities

Artificial Intelligence, In vivo models, Ex vivo screens and global clinical genomics

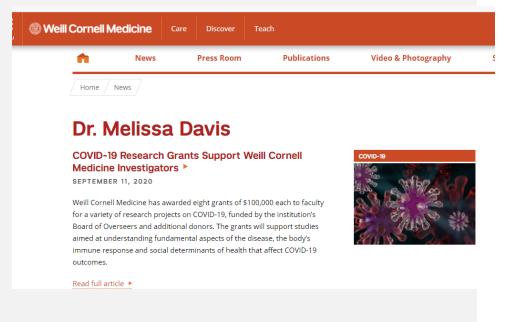


DARC STARR Premise : DARC-regulated Race-group TME differences

*Obesity-related? *Ancestry-related? *Duffy-null(Fy-)-regulated?

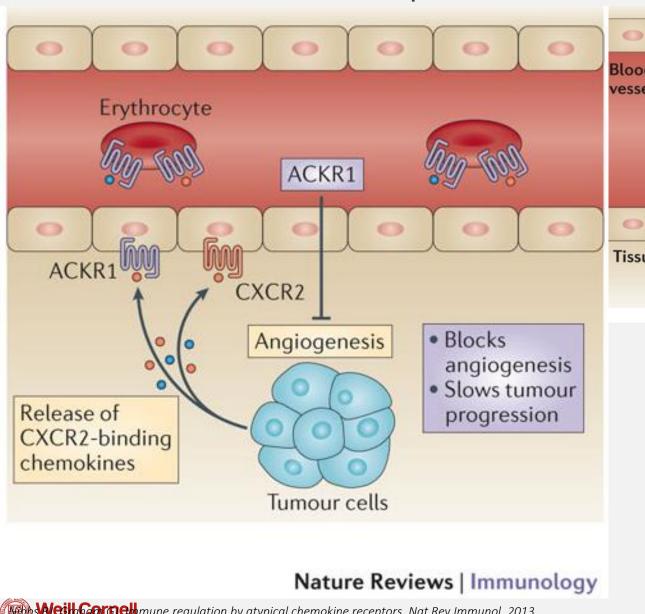


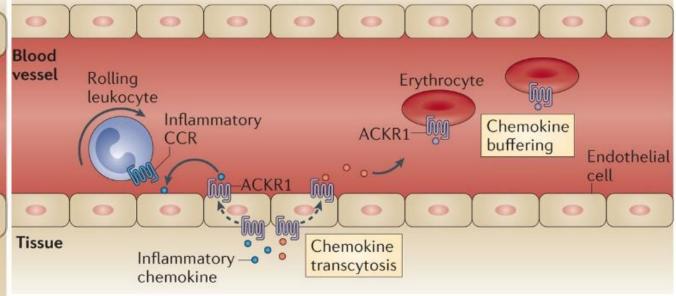




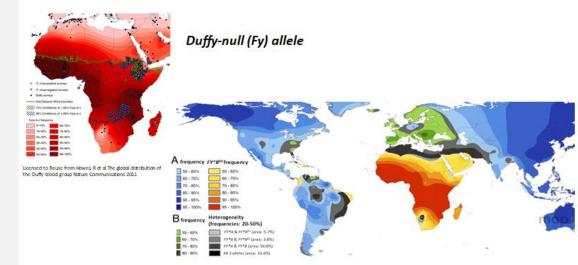


DARC regulates dynamic inflammatory chemokine levels via vascular endothelial cells and concentrations in plasma



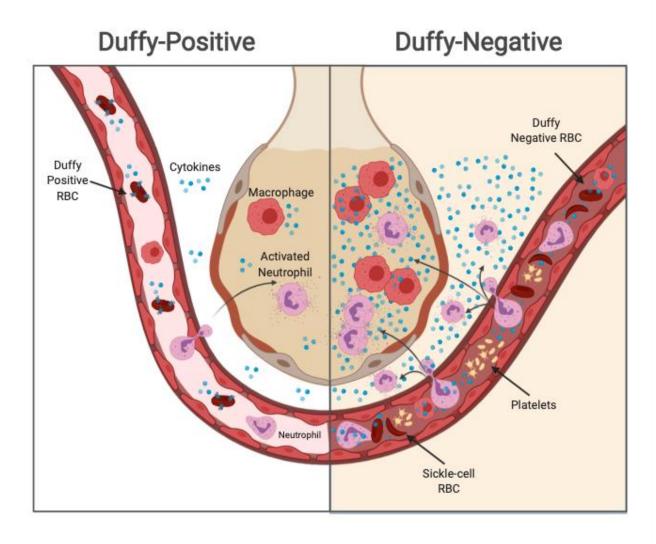


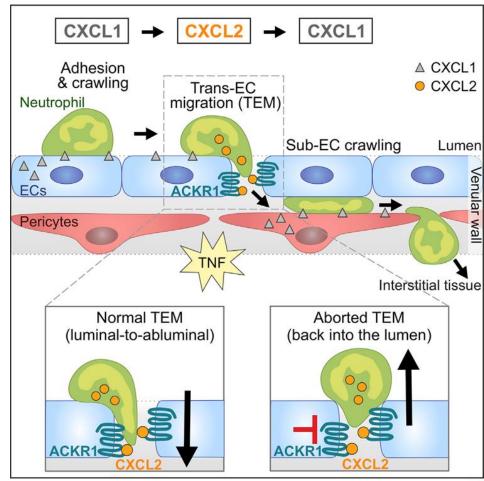
Genetic imprint of the West African Diaspora Impact on TNBC Risk?



Weill Cornell mune regulation by atypical chemokine receptors. Nat Rev Immunol. 2013 Medicine

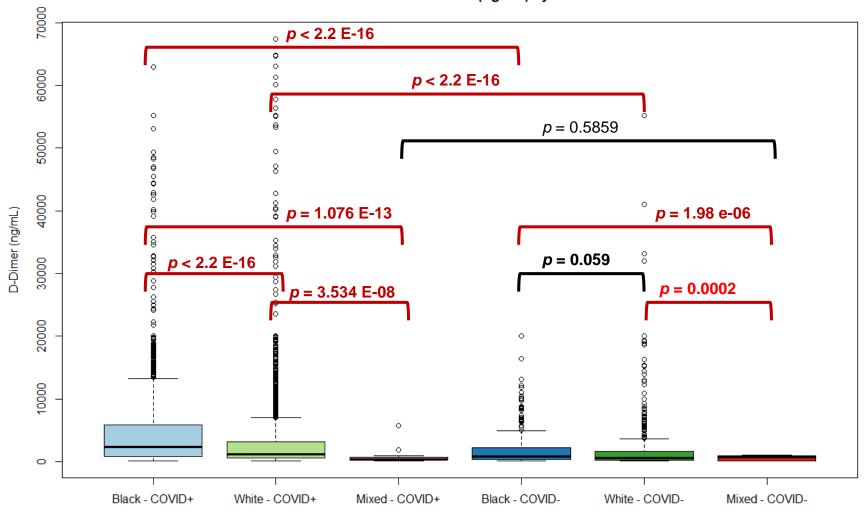
The DARC side of COVID-19 Disparities





Breast Cancer Subtypes

Box Plots of D-Dimer (ng/mL) by Race

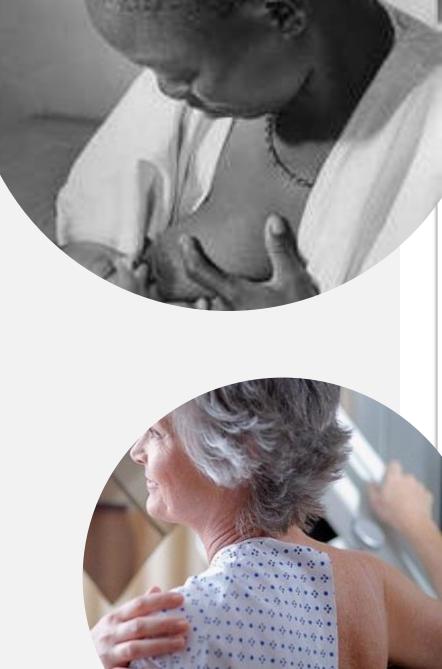






Summary

- Disparities in cancer mortality can be driven by biological distinctions
- INCLUSION of genetically diverse populations will reveal distinct mechanisms and/or functionality of genes involved in tumor progression
- Tumor immune response may drive race/ancestry group differences due to evolutionary adaptation across populations
- DARC tumor phenotypes have more infiltrating immune cells and distinct cell markers within immune cell groups
- Precision medicine tools can help identify new ways of treating disease phenotypes that are biased to certain populations





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